

# HIGH-ORDER SOLUTION TRANSFER BETWEEN CURVED TRIANGULAR MESHES

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**ABSTRACT.** The problem of solution transfer between meshes arises frequently in computational physics, e.g., in Lagrangian methods where remeshing occurs. The interpolation process must be conservative, i.e., it must conserve physical properties, such as mass. We extend previous works — which described the solution transfer process for straight sided unstructured meshes — by considering high-order isoparametric meshes with curved elements. To facilitate solution transfer, we numerically integrate the product of shape functions via Green’s theorem along the boundary of the intersection of two curved elements. We perform a numerical experiment and confirm the expected accuracy by transferring test fields across two families of meshes.

*Keywords:* Remapping, Curved Meshes, Lagrangian, Solution Transfer, Discontinuous Galerkin

## 1. INTRODUCTION

Solution transfer between meshes is a common problem that has practical applications to many methods in computational physics. For example, by allowing the underlying computational domain to change during a simulation, computational effort can be focused dynamically to resolve relevant features of a numerical solution. This so-called mesh adaptivity typically requires translating the numerical solution from the old mesh to the new, that is, solution transfer [BR78, PVMZ87, PUdOG01, IK04]. As another example, Lagrangian or particle-based methods treat each node in the mesh as a particle and so with each timestep the mesh travels *with* the fluid [DKR12]. However, over (typically short) time the mesh becomes distorted and suffers a loss in element quality which eventually causes inaccurate solutions. To overcome this, the domain must be remeshed or rezoned and the solution must be transferred (remapped) onto the new mesh configuration. A related more general class of methods is the Arbitrary Lagrangian-Eulerian (ALE) method, which are not constrained by a fixed computational mesh (the Eulerian approach) or by a fixed fluid flow (the Lagrangian approach) and typically combine the benefits of both approaches [HAC74, ADKR15].

Solution transfer is needed when a solution (approximated by a discrete field) is known on a *donor* mesh and must be transferred to a *target* mesh. When pointwise interpolation is used to transfer a solution, quantities with physical meaning (for instance, mass, concentration, energy) may not be conserved. In many applications, the field must be conserved for physical reasons, e.g., mass or energy cannot leave or enter the system; hence we focus on *conservative* solution transfer (typically using Galerkin or  $L_2$ -minimizing methods).

The problem of conservative interpolation was considered previously for straight sided meshes. However, both to allow for greater geometric flexibility and for high order of convergence, we address the case of curved isoparametric meshes. This is based on the recent interest in so-called high-order methods [WFA<sup>+</sup>13], which have the ability to produce highly accurate solutions with low dissipation and low dispersion error. However, these methods typically require curved meshes to obtain the high accuracy.

The *common refinement* approach in [JH04] is used to compare several methods for solution transfer across two meshes. The problem of constructing such a refinement is considered in [FPP<sup>+</sup>09, FM11] (called

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a supermesh by the authors). However, the solution transfer becomes considerably more challenging for curved meshes. For a sense of the difference between the straight sided and curved cases, consider the problem of intersecting an element from the donor mesh with an element from the target mesh. If the elements are triangles, the intersection is either a convex polygon or has measure zero. If the elements are curved, the intersection can be non-convex and can even split into multiple disjoint regions. In [CGQ18], a related problem is considered where squares are allowed to curve along characteristics and the solution is remapped onto a regular square grid.

Conservative solution transfer has been around since the advent of ALE, and as a result much of the existing literature focuses on mesh-mesh pairs that will occur during an ALE-based simulation. Many ALE methods modify the mesh during simulation: when flow-based mesh distortion occurs, elements are typically “flipped” (for example, a diagonal is switched in a pair of elements) or elements are subdivided or combined. These operations are inherently local; hence the solution transfer can be done locally across known neighbors. Typically, this locality is crucial to solution transfer methods. In [MS03], the transfer is based on partitioning elements of the updated mesh into components of elements from the old mesh and “swept regions” from neighboring elements. In [KS08], the (locally) changing connectivity of the mesh is addressed. In [GKS07], the local transfer is done on polyhedral meshes.

Global solution transfer instead seeks to conserve the solution across the whole mesh. It makes no assumptions about the relationship between the donor and target meshes. The loss in local information makes the mesh intersection problem more computationally expensive, but the added flexibility reduces timestep restrictions since it allows remeshing to be done less often. In [Duk84, DK87], a global transfer is enabled by transforming volume integrals to surface integrals via the divergence theorem to reduce the complexity of the problem.

In addition to conservation, there are many other considerations for a solution transfer method that are important in various applications. For example, since our focus is on high-order methods, these typically produce oscillations in the solutions that can be non-physical and break the numerical solvers. This can happen for the standard Galerkin projections that we consider here, which means that e.g., discontinuities such as shocks in the solution could become nonphysical after our solution transfer. Various solutions to this have been proposed before, often referred to as limiting or bound/monotonicity preserving methods; see, for instance, [ADKR15, ADK<sup>+</sup>18] where several approaches are studied including a high-order extension of the flux-corrected transport (FCT) method. Other considerations for a solution transfer scheme include the preservation of additional constraints, such as the divergence or the curl of the solution.

We describe an algorithm for conservative solution transfer between curved meshes will be described. This method applies to meshes in  $\mathbf{R}^2$ . Application to meshes in  $\mathbf{R}^3$  is a direction for future research, though the geometric kernels (see Chapter A) become significantly more challenging to describe and implement. In addition, the method will assume that every element in the target mesh is contained in the donor mesh. This ensures that the solution transfer is *interpolative*. In the case where all target elements are partially covered, *extrapolation* could be used to extend a solution outside the domain, but for totally uncovered elements there is no clear correspondence to elements in the donor mesh.

This paper is organized as follows. Section 2 establishes common notation and reviews basic results relevant to the topics at hand. Section A is an in-depth discussion of the computational geometry methods needed to implement to enable solution transfer. Section 3 (and following) describes the solution transfer process and gives results of some numerical experiments confirming the rate of convergence.

## 2. PRELIMINARIES

**2.1. General Notation.** We’ll refer to  $\mathbf{R}$  for the reals,  $\mathcal{U}$  represents the unit triangle (or unit simplex) in  $\mathbf{R}^2$ :

$$\mathcal{U} = \{(s, t) \mid 0 \leq s, t, s + t \leq 1\}. \quad (2.1)$$

When dealing with sequences with multiple indices, for example,  $s_{m,n} = m + n$ , we’ll use bold symbols to represent a multi-index:  $\mathbf{i} = (m, n)$ . The binomial coefficient  $\binom{n}{k}$  is equal to  $\frac{n!}{k!(n-k)!}$  and the trinomial coefficient  $\binom{n}{i,j,k}$  is equal to  $\frac{n!}{i!j!k!}$  (where  $i + j + k = n$ ). The notation  $\delta_{ij}$  represents the Kronecker delta, a value which is 1 when  $i = j$  and 0 otherwise.

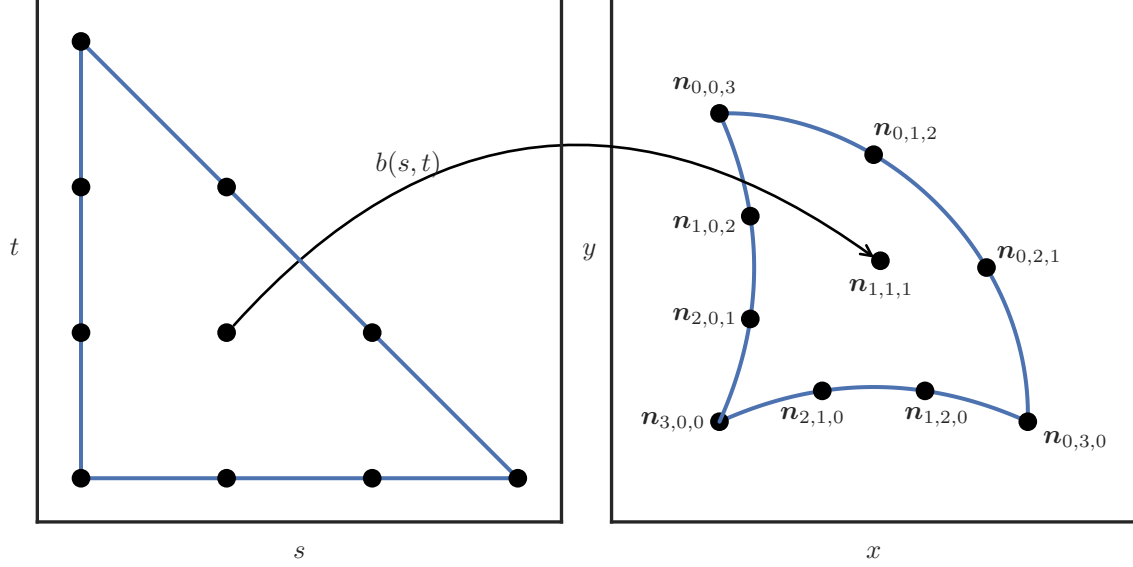


Figure 2.1. Cubic Bézier triangle

**2.2. Bézier Curves.** A *Bézier curve* is a mapping from the unit interval that is determined by a set of control points  $\{\mathbf{p}_j\}_{j=0}^n \subset \mathbf{R}^d$ . For a parameter  $s \in [0, 1]$ , there is a corresponding point on the curve:

$$b(s) = \sum_{j=0}^n \binom{n}{j} (1-s)^{n-j} s^j \mathbf{p}_j \in \mathbf{R}^d. \quad (2.2)$$

This is a combination of the control points weighted by each Bernstein basis function  $B_{j,n}(s) = \binom{n}{j} (1-s)^{n-j} s^j$ . Due to the binomial expansion  $1 = (s + (1-s))^n = \sum_{j=0}^n B_{j,n}(s)$ , a Bernstein basis function is in  $[0, 1]$  when  $s$  is as well. Due to this fact, the curve must be contained in the convex hull of its control points.

**2.3. Bézier Triangles.** A *Bézier triangle* [Far01, Chapter 17] is a mapping from the unit triangle  $\mathcal{U}$  and is determined by a control net  $\{\mathbf{p}_{i,j,k}\}_{i+j+k=n} \subset \mathbf{R}^d$ . A Bézier triangle is a particular kind of Bézier surface, i.e., one in which there are two cartesian or three barycentric input parameters. Often the term Bézier surface is used to refer to a tensor product or rectangular patch. For  $(s, t) \in \mathcal{U}$  we can define barycentric weights  $\lambda_1 = 1 - s - t$ ,  $\lambda_2 = s$ ,  $\lambda_3 = t$  so that

$$1 = (\lambda_1 + \lambda_2 + \lambda_3)^n = \sum_{\substack{i+j+k=n \\ i,j,k \geq 0}} \binom{n}{i,j,k} \lambda_1^i \lambda_2^j \lambda_3^k. \quad (2.3)$$

Using this we can similarly define a (triangular) Bernstein basis

$$B_{i,j,k}(s, t) = \binom{n}{i,j,k} (1-s-t)^i s^j t^k = \binom{n}{i,j,k} \lambda_1^i \lambda_2^j \lambda_3^k \quad (2.4)$$

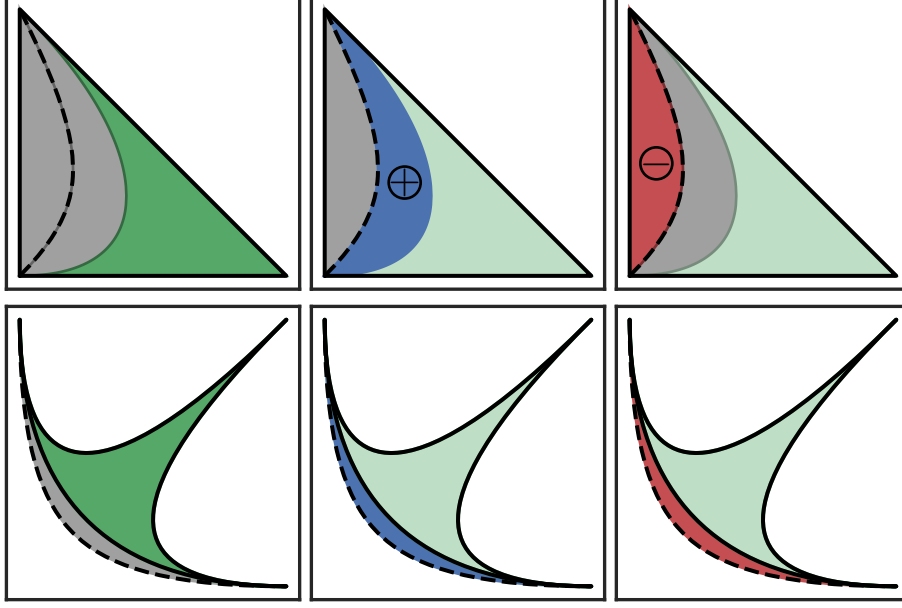
that is in  $[0, 1]$  when  $(s, t)$  is in  $\mathcal{U}$ . Using this, we define points on the Bézier triangle as a convex combination of the control net:

$$b(s, t) = \sum_{i+j+k=n} \binom{n}{i,j,k} \lambda_1^i \lambda_2^j \lambda_3^k \mathbf{p}_{i,j,k} \in \mathbf{R}^d. \quad (2.5)$$

Rather than defining a Bézier triangle by the control net, it can also be uniquely determined by the image of a standard lattice of points in  $\mathcal{U}$ :  $b(j/n, k/n) = \mathbf{n}_{i,j,k}$ ; we'll refer to these as *standard nodes*. Figure 2.1 shows these standard nodes form a cubic triangle in  $\mathbf{R}^2$ . To see the correspondence, when  $p = 1$  the standard nodes are the control net

$$b(s, t) = \lambda_1 \mathbf{n}_{1,0,0} + \lambda_2 \mathbf{n}_{0,1,0} + \lambda_3 \mathbf{n}_{0,0,1} \quad (2.6)$$

and when  $p = 2$



**Figure 2.2.** The Bézier triangle given by  $b(s, t) = [(1 - s - t)^2 + s^2 \quad s^2 + t^2]^T$  produces an inverted element. It traces the same region twice, once with a positive Jacobian (the middle column) and once with a negative Jacobian (the right column).

$$b(s, t) = \lambda_1 (2\lambda_1 - 1) \mathbf{n}_{2,0,0} + \lambda_2 (2\lambda_2 - 1) \mathbf{n}_{0,2,0} + \lambda_3 (2\lambda_3 - 1) \mathbf{n}_{0,0,2} + 4\lambda_1 \lambda_2 \mathbf{n}_{1,1,0} + 4\lambda_2 \lambda_3 \mathbf{n}_{0,1,1} + 4\lambda_3 \lambda_1 \mathbf{n}_{1,0,1}. \quad (2.7)$$

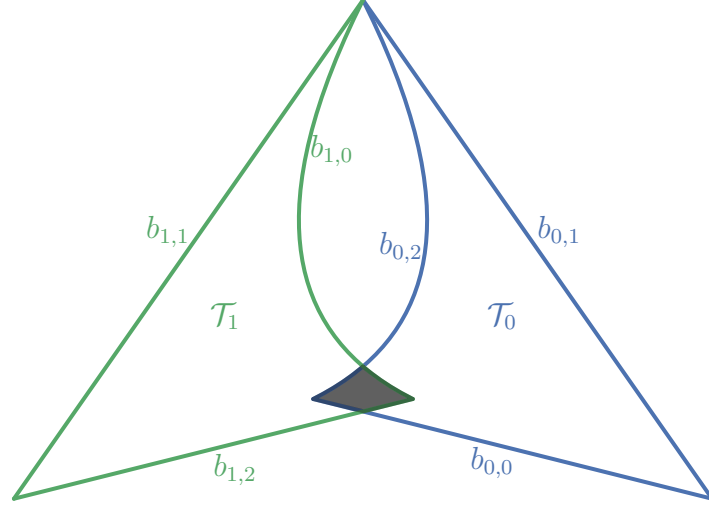
However, it's worth noting that the transformation between the control net and the standard nodes has condition number that grows exponentially with  $n$  (see [Far91], which is related but does not directly show this). This may make working with higher-degree triangles prohibitively unstable.

A *valid* Bézier triangle is one which is diffeomorphic to  $\mathcal{U}$ , that is,  $b(s, t)$  is bijective and has an invertible Jacobian everywhere. We must also have the orientation preserved, that is, the Jacobian must have positive determinant. For example, in Figure 2.2, the image of  $\mathcal{U}$  under the map  $b(s, t) = [(1 - s - t)^2 + s^2 \quad s^2 + t^2]^T$  is not valid because the Jacobian is zero along the curve  $s^2 - st - t^2 - s + t = 0$  (the dashed line). Elements that are not valid are called *inverted* because they have regions with “negative area”. For the example, the image  $b(\mathcal{U})$  leaves the boundary determined by the edge curves:  $b(r, 0)$ ,  $b(1 - r, r)$  and  $b(0, 1 - r)$  when  $r \in [0, 1]$ . This region outside the boundary is traced twice, once with a positive Jacobian and once with a negative Jacobian.

**2.4. Curved Elements.** We define a curved mesh element  $\mathcal{T}$  of degree  $p$  to be a Bézier triangle in  $\mathbf{R}^2$  of the same degree. We refer to the component functions of  $b(s, t)$  (the map that gives  $\mathcal{T} = b(\mathcal{U})$ ) as  $x(s, t)$  and  $y(s, t)$ .

This fits a typical definition ([JM09, Chapter 12]) of a curved element, but gives a special meaning to the mapping from the reference triangle. Interpreting elements as Bézier triangles has been used for Lagrangian methods where mesh adaptivity is needed (e.g., [CMOP04]). Typically curved elements only have one curved side ([MM72]) since they are used to resolve geometric features of a boundary. See also [Zlá73, Zlá74]. Bézier curves and triangles have a number of mathematical properties (e.g., the convex hull property) that lead to elegant geometric descriptions and algorithms.

Note that a Bézier triangle can be determined from many different sources of data (for example, the control net or the standard nodes). The choice of this data may be changed to suit the underlying physical problem without changing the actual mapping. Conversely, the data can be fixed (for instance, as the control net) to avoid costly basis conversion; once fixed, the equations of motion and other PDE terms can be recast



**Figure 2.3.** Intersection of Bézier triangles form a curved polygon.

relative to the new basis (for an example, see [PBP09], where the domain varies with time but the problem is reduced to solving a transformed conservation law in a fixed reference configuration).

**2.5. Shape Functions.** When defining shape functions (i.e., a basis with geometric meaning) on a curved element there are (at least) two choices. When the degree of the shape functions is the same as the degree of the function being represented on the Bézier triangle, we say the element  $\mathcal{T}$  is *isoparametric*. For the multi-index  $\mathbf{i} = (i, j, k)$ , we define  $\mathbf{u}_i = (j/n, k/n)$  and the corresponding standard node  $\mathbf{n}_i = b(\mathbf{u}_i)$ . Given these points, two choices for shape functions present themselves:

- *preimage basis*:  $\phi_j(\mathbf{n}_i) = \hat{\phi}_j(\mathbf{u}_i) = \hat{\phi}_j(b^{-1}(\mathbf{n}_i))$  where  $\hat{\phi}_j$  is a canonical basis function on  $\mathcal{U}$ , that is,  $\hat{\phi}_j$  is a degree  $p$  bivariate polynomial and  $\hat{\phi}_j(\mathbf{u}_i) = \delta_{ij}$ .
- *global coordinates basis*:  $\phi_j(\mathbf{n}_i) = \delta_{ij}$ , that is, a canonical basis function on the standard nodes  $\{\mathbf{n}_i\}$ .

For example, consider a quadratic Bézier triangle:

$$b(s, t) = \begin{bmatrix} 4(st + s + t) & 4(st + t + 1) \end{bmatrix}^T \quad (2.8)$$

$$\implies \begin{bmatrix} \mathbf{n}_{2,0,0} & \mathbf{n}_{1,1,0} & \mathbf{n}_{0,2,0} & \mathbf{n}_{1,0,1} & \mathbf{n}_{0,1,1} & \mathbf{n}_{0,0,2} \end{bmatrix} = \begin{bmatrix} 0 & 2 & 4 & 2 & 5 & 4 \\ 4 & 4 & 4 & 6 & 7 & 8 \end{bmatrix}. \quad (2.9)$$

In the *global coordinates basis*, we have

$$\phi_{0,1,1}^G(x, y) = \frac{(y-4)(x-y+4)}{6}. \quad (2.10)$$

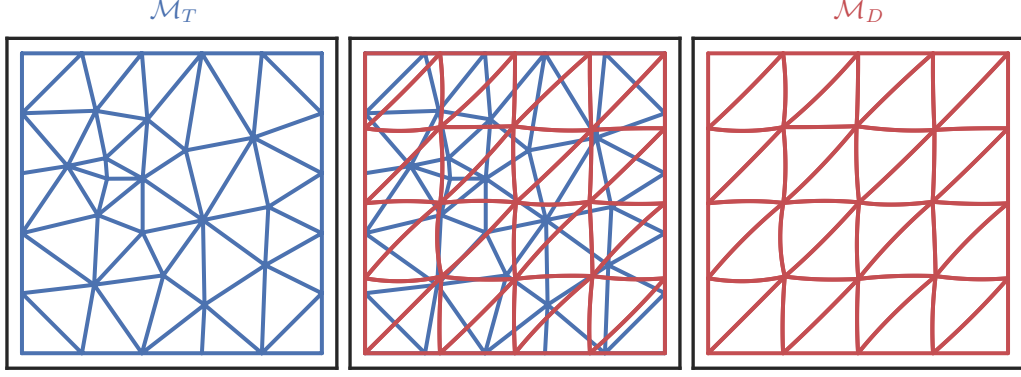
For the *preimage basis*, we need the inverse and the canonical basis

$$b^{-1}(x, y) = \begin{bmatrix} \frac{x-y+4}{4} & \frac{y-4}{x-y+8} \end{bmatrix} \quad \text{and} \quad \hat{\phi}_{0,1,1}(s, t) = 4st, \quad (2.11)$$

and together they give

$$\phi_{0,1,1}^P(x, y) = \frac{(y-4)(x-y+4)}{x-y+8}. \quad (2.12)$$

In general  $\phi_j^P$  may not even be a rational bivariate function; due to composition with  $b^{-1}$  we can only guarantee that it is algebraic (i.e., it can be defined as the zero set of polynomials with coefficients in  $\mathbf{R}[x, y]$ ).



**Figure 3.1.** Mesh pair: donor mesh  $\mathcal{M}_D$  and target mesh  $\mathcal{M}_T$ .

**2.6. Curved Polygons.** When intersecting two curved elements, the resulting surface(s) will be defined by the boundary, alternating between edges of each element. For example, in Figure 2.3, a “curved quadrilateral” is formed when two Bézier triangles  $\mathcal{T}_0$  and  $\mathcal{T}_1$  are intersected.

A *curved polygon* is defined by a collection of Bézier curves in  $\mathbf{R}^2$  that determine the boundary. In order to be a valid polygon, none of the boundary curves may cross, the ends of consecutive edge curves must meet and the curves must be right-hand oriented. For our example in Figure 2.3, the triangles have boundaries formed by three Bézier curves:  $\partial\mathcal{T}_0 = b_{0,0} \cup b_{0,1} \cup b_{0,2}$  and  $\partial\mathcal{T}_1 = b_{1,0} \cup b_{1,1} \cup b_{1,2}$ . The intersection  $\mathcal{P}$  is defined by four boundary curves:  $\partial\mathcal{P} = C_1 \cup C_2 \cup C_3 \cup C_4$ . Each boundary curve is itself a Bézier curve:<sup>1</sup>  $C_1 = b_{0,0}([0, 1/8])$ ,  $C_2 = b_{1,2}([7/8, 1])$ ,  $C_3 = b_{1,0}([0, 1/7])$  and  $C_4 = b_{0,2}([6/7, 1])$ .

Though an intersection can be described in terms of the Bézier triangles, the structure of the control net will be lost. The region will not in general be able to be described by a mapping from a simple space like  $\mathcal{U}$ .

### 3. GALERKIN PROJECTION

**3.1. Linear System for Projection.** Consider a donor mesh  $\mathcal{M}_D$  with shape function basis  $\phi_D^{(j)}$  and a known field <sup>2</sup>  $\mathbf{q}_D = \sum_j d_j \phi_D^{(j)}$  and a target mesh  $\mathcal{M}_T$  with shape function basis  $\phi_T^{(j)}$  (Figure 3.1). Each shape function  $\phi$  corresponds to a given isoparametric curved element (see Section 2.4)  $\mathcal{T}$  in one of these meshes and has  $\text{supp}(\phi) = \mathcal{T}$ . Additionally, the shape functions are degree- $p$  polynomials (see Section 2.5 for a discussion of shape functions), but the degree of the donor mesh need not be the same as that of the target mesh. We assume that both meshes cover the same domain  $\Omega \subset \mathbf{R}^2$ , however we really only require the donor mesh to cover the target mesh.

We seek the  $L_2$ -optimal interpolant  $\mathbf{q}_T = \sum_j t_j \phi_T^{(j)}$ :

$$\|\mathbf{q}_T - \mathbf{q}_D\|_2 = \min_{\mathbf{q} \in \mathcal{V}_T} \|\mathbf{q} - \mathbf{q}_D\|_2, \quad (3.1)$$

where  $\mathcal{V}_T = \text{Span}_j \{\phi_T^{(j)}\}$  is the function space defined on the target mesh. Since this is optimal in the  $L_2$  sense, by differentiating with respect to each  $t_j$  in  $\mathbf{q}_T$  we find the weak form:

$$\int_{\Omega} \mathbf{q}_D \phi_T^{(j)} dV = \int_{\Omega} \mathbf{q}_T \phi_T^{(j)} dV \quad \text{for all } j. \quad (3.2)$$

If the constant function 1 is contained in  $\mathcal{V}_T$ , conservation follows from the weak form and linearity of the integral

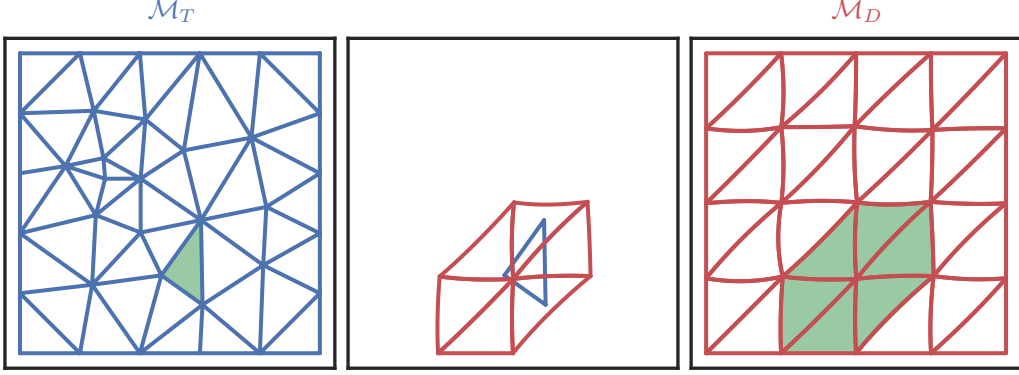
$$\int_{\Omega} \mathbf{q}_D dV = \int_{\Omega} \mathbf{q}_T dV. \quad (3.3)$$

Expanding  $\mathbf{q}_D$  and  $\mathbf{q}_T$  with respect to their coefficients  $\mathbf{d}$  and  $\mathbf{t}$ , the weak form gives rise to a linear system

$$M_T \mathbf{t} = M_{TD} \mathbf{d}. \quad (3.4)$$

<sup>1</sup>A specialization of a Bézier curve  $b([a_1, a_2])$  is also a Bézier curve.

<sup>2</sup>This is somewhat a simplification. In CG (continuous Galerkin, as opposed to DG or discontinuous Galerkin), some coefficients will be paired with multiple shape functions, such as the coefficient at a vertex node.



**Figure 3.2.** All donor elements that cover a target element.

Here  $M_T$  is the mass matrix for  $\mathcal{M}_T$  given by

$$(M_T)_{ij} = \int_{\Omega} \phi_T^{(i)} \phi_T^{(j)} dV. \quad (3.5)$$

In the discontinuous Galerkin case,  $M_T$  is block diagonal with blocks that correspond to each element, so (3.4) can be solved locally on each element  $\mathcal{T}$  in the target mesh. By construction,  $M_T$  is symmetric and sparse since  $(M_T)_{ij}$  will be 0 unless  $\phi_T^{(i)}$  and  $\phi_T^{(j)}$  are supported on the same element  $\mathcal{T}$ . In the continuous case,  $M_T$  is globally coupled since coefficients corresponding to boundary nodes interact with multiple elements. The matrix  $M_{TD}$  is a “mixed” mass matrix between the target and donor meshes:

$$(M_{TD})_{ij} = \int_{\Omega} \phi_T^{(i)} \phi_D^{(j)} dV. \quad (3.6)$$

Boundary conditions can be imposed on the system by fixing some coefficients, but that is equivalent to removing some of the basis functions which may in turn make the projection non-conservative. This is because the removed basis functions may have been used in  $1 = \sum_j u_j \phi_T^{(j)}$ .

Computing  $M_T$  is fairly straightforward since the (bidirectional) mapping from elements  $\mathcal{T}$  to basis functions  $\phi_T^{(j)}$  supported on those elements is known. When using shape functions in the global coordinates basis (see Section 2.5), the integrand  $F = \phi_T^{(i)} \phi_T^{(j)}$  will be a polynomial of degree  $2p$  on  $\mathbf{R}^2$ . The domain of integration  $\mathcal{T} = b(\mathcal{U})$  is the image of a (degree- $p$ ) map  $b(s, t)$  from the unit triangle. Using substitution

$$\int_{b(\mathcal{U})} F(x, y) dx dy = \int_{\mathcal{U}} \det(Db) F(x(s, t), y(s, t)) ds dt \quad (3.7)$$

(we know the map preserves orientation, i.e.  $\det(Db)$  is positive). Once transformed this way, a quadrature rule on the unit triangle ([Dun85]) can be used.

On the other hand, computing  $M_{TD}$  is significantly more challenging. This requires solving both a geometric problem — finding the region to integrate over — and an analytic problem — computing the integrals. The integration can be done with a quadrature rule, though finding this region is significantly more difficult.

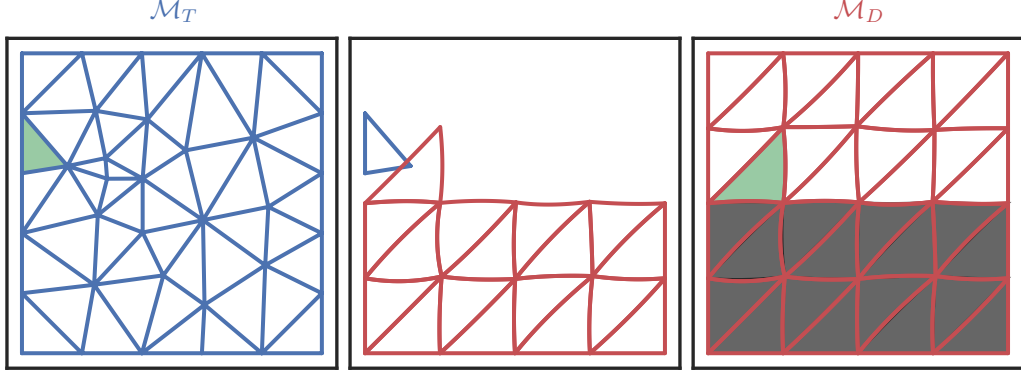
**3.2. Common Refinement.** Rather than computing  $M_{TD}$ , the right-hand side of (3.4) can be computed directly via

$$(M_{TD}\mathbf{d})_j = \int_{\Omega} \phi_T^{(j)} \mathbf{q}_D dV. \quad (3.8)$$

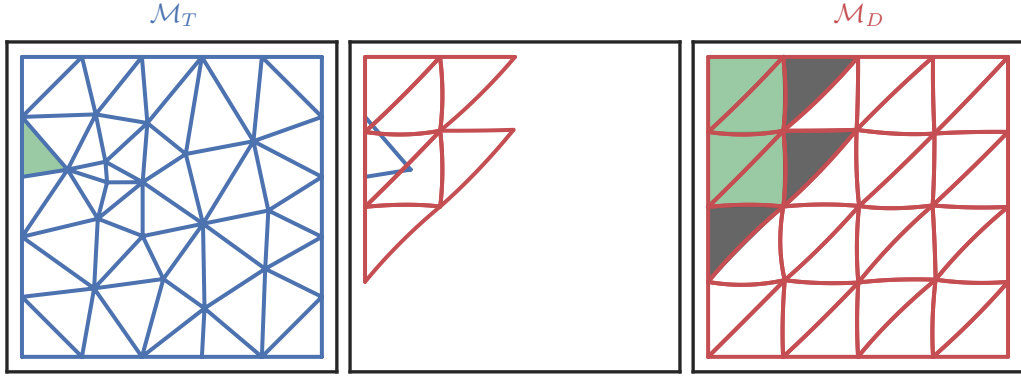
Any given  $\phi$  is supported on an element  $\mathcal{T}$  in the target mesh. Since  $\mathbf{q}_D$  is piecewise defined over each element  $\mathcal{T}'$  in the donor mesh, the integral (3.8) may be problematic. In the continuous Galerkin case,  $\mathbf{q}_D$  need not be differentiable across  $\mathcal{T}$  and in the discontinuous Galerkin case,  $\mathbf{q}_D$  need not even be continuous. This necessitates a partitioning of the domain:

$$\int_{\Omega} \phi \mathbf{q}_D dV = \int_{\mathcal{T}} \phi \mathbf{q}_D dV = \sum_{\mathcal{T}' \in \mathcal{M}_D} \int_{\mathcal{T} \cap \mathcal{T}'} \phi \mathbf{q}_D|_{\mathcal{T}'} dV. \quad (3.9)$$





**Figure 3.3.** Brute force search for a donor element  $\mathcal{T}'$  that matches a fixed target element  $\mathcal{T}$ .



**Figure 3.4.** All donor elements  $\mathcal{T}'$  that cover a target element  $\mathcal{T}$ , with an extra layer of neighbors in the donor mesh that *do not* intersect  $\mathcal{T}$ .

In other words, the integral over  $\mathcal{T}$  splits into integrals over intersections  $\mathcal{T} \cap \mathcal{T}'$  for all  $\mathcal{T}'$  in the donor mesh that intersect  $\mathcal{T}$  (Figure 3.2). Since both  $\phi$  and  $\mathbf{q}_D|_{\mathcal{T}'}$  are polynomials on  $\mathcal{T} \cap \mathcal{T}'$ , the integrals will be exact when using a quadrature scheme of an appropriate degree of accuracy. Without partitioning  $\mathcal{T}$ , the integrand is not a polynomial (in fact, possibly not smooth), so the quadrature cannot be exact.

In order to compute  $M_{TD}\mathbf{d}$ , we'll need to compute the *common refinement*, that is, an intermediate mesh that contains both the donor and target meshes. This will consist of all non-empty  $\mathcal{T} \cap \mathcal{T}'$  as  $\mathcal{T}$  varies over elements of the target mesh and  $\mathcal{T}'$  over elements of the donor mesh. This requires solving three specific subproblems:

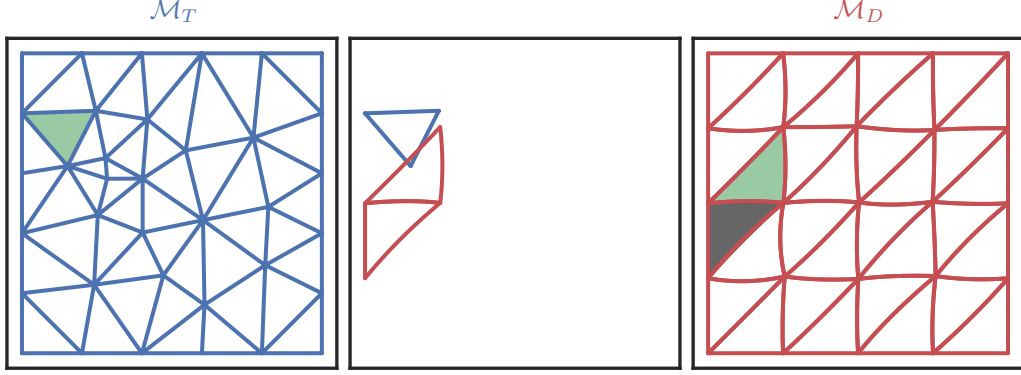
- Forming the region(s) of intersection between two elements that are Bézier triangles.
- Finding all pairs of elements, one each from the target and donor mesh, that intersect.
- Numerically integrating over a region of intersection between two elements.

The first subproblem is discussed in Section A.2 and the curved polygon region(s) of intersection have been described in Section 2.6. The second will be considered in Section 3.3 and the third in Section 3.4 below.

**3.3. Expanding Front.** We seek to identify all pairs  $\mathcal{T}$  and  $\mathcal{T}'$  of intersecting target and donor elements. The naïve approach just considers every pair of elements and takes  $\mathcal{O}(|\mathcal{M}_D||\mathcal{M}_T|)$  to complete.<sup>3</sup> Taking after [FM11], we can do much better than this quadratic time search. In fact, we can compute all integrals in  $\mathcal{O}(|\mathcal{M}_D| + |\mathcal{M}_T|)$ . First, we fix an element of the target mesh and perform a brute-force search to find an intersecting element in the donor mesh (Figure 3.3). This has worst-case time  $\mathcal{O}(|\mathcal{M}_D|)$ .

<sup>3</sup>For a mesh  $\mathcal{M}$ , the expression  $|\mathcal{M}|$  represents the number of elements in the mesh.





**Figure 3.5.** First match between a neighbor of the previously considered target element and all the donor elements that match the previously considered target element.

Once we have such a match, we use the connectivity graph of the donor mesh to perform a breadth-first search for neighbors that also intersect the target element  $\mathcal{T}$  (Figure 3.4). This search takes  $\mathcal{O}(1)$  time. It's also worthwhile to keep the first layer of donor elements that don't intersect  $\mathcal{T}$  because they are more likely to intersect the neighbors of  $\mathcal{T}$ .<sup>4</sup> Using the list of intersected elements, a neighbor of  $\mathcal{T}$  can find a donor element it intersects with in  $\mathcal{O}(1)$  time (Figure 3.5). As seen, after our  $\mathcal{O}(|\mathcal{M}_D|)$  initial brute-force search, the localized intersections for each target element  $\mathcal{T}$  take  $\mathcal{O}(1)$  time. So together, the process takes  $\mathcal{O}(|\mathcal{M}_D| + |\mathcal{M}_T|)$ .

For special cases such as ALE methods, the initial brute-force search can be reduced to  $\mathcal{O}(1)$ . This could be enabled by tracking the remeshing process so a correspondence already exists. If a full mapping from donor to target mesh exists, the process of computing  $M_T$  and  $M_{TD}\mathbf{d}$  can be fully parallelized across elements of the target mesh, or even across shape functions  $\phi_T^{(j)}$ . Spatial binning techniques for mesh elements could also be used to improve the initial brute-force search as in [WZRC16].

**3.4. Integration over Curved Polygons.** In order to numerically evaluate integrals of the form

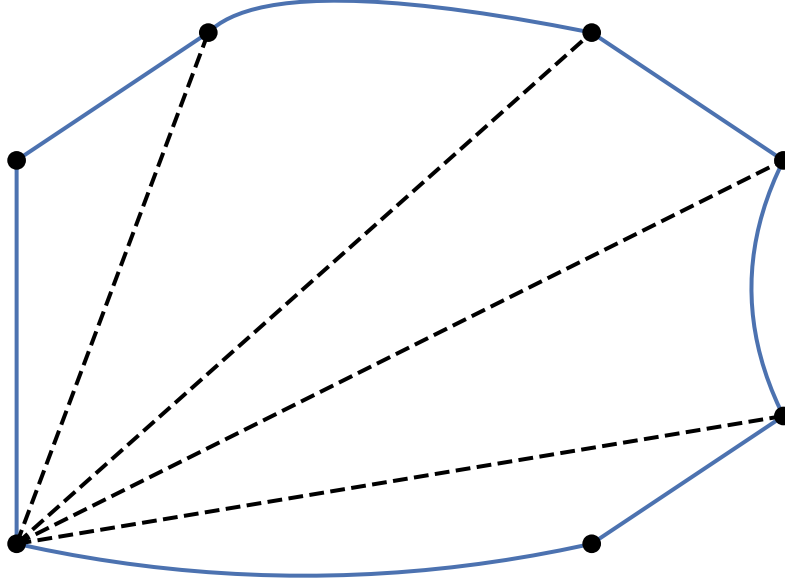
$$\int_{\mathcal{T}_0 \cap \mathcal{T}_1} F(x, y) dV \quad (3.10)$$

we must have a quadrature rule on these curved polygon (Section 2.6) intersections  $\mathcal{T}_0 \cap \mathcal{T}_1$ . To do this, we transform the integral into several line integrals via Green's theorem and then use an exact Gaussian quadrature to evaluate them. Throughout this section we'll assume the integrand is of the form  $F = \phi_0 \phi_1$  where each  $\phi_j$  is a shape function on  $\mathcal{T}_j$ . In addition, we'll refer to the two Bézier maps that define the elements being intersected:  $\mathcal{T}_0 = b_0(\mathcal{U})$  and  $\mathcal{T}_1 = b_1(\mathcal{U})$ .

First a discussion of a method not used. A somewhat simple approach would be to use polygonal approximation, that is, approximate the boundary of each Bézier triangle with line segments, intersect the resulting polygons, triangulate the intersection polygon(s) and then numerically integrate on each triangulated cell. However, this approach is prohibitively inefficient. For example, consider computing the area of  $\mathcal{P}$  via an integral:  $\int_{\mathcal{P}} 1 dV$ . By using the actual curved boundaries of each element, this integral can be computed with relative error on the order of machine precision  $\mathcal{O}(\mathbf{u})$ . On the other hand, approximating each side of  $\mathcal{P}$  with  $N$  line segments, the relative error is  $\mathcal{O}(1/N^2)$ . (For example, if  $N = 2$  an edge curve  $b(s, 0)$  would be replaced by segments connecting  $b(0, 0)$ ,  $b(1/2, 0)$  and  $b(1, 0)$ , i.e.,  $N + 1$  equally spaced parameters.) This means that in order to perform as well as an *exact* quadrature used in the curved case, we'd need  $N = \mathcal{O}(1/\sqrt{\mathbf{u}})$ .

Since polygonal approximation is prohibitively expensive, we work directly with curved edges and compute integrals on a regular domain via substitution. If two Bézier triangles intersect with positive measure, then

<sup>4</sup>In the very unlikely case that the boundary of  $\mathcal{T}$  *exactly matches* the boundaries of the donor elements that cover it, *none* of the overlapping donor elements can intersect the neighbors of  $\mathcal{T}$  so the first layer of non-intersected donor elements must be considered.



**Figure 3.6.** Curved polygon tessellation, done by introducing diagonals from a single vertex node.

the region of intersection is one or more disjoint curved polygons:  $\mathcal{T}_0 \cap \mathcal{T}_1 = \mathcal{P}$  or  $\mathcal{T}_0 \cap \mathcal{T}_1 = \mathcal{P} \cup \mathcal{P}' \cup \dots$ . The second case can be handled in the same way as the first by handling each disjoint region independently:

$$\int_{\mathcal{T}_0 \cap \mathcal{T}_1} F(x, y) dV = \int_{\mathcal{P}} F(x, y) dV + \int_{\mathcal{P}'} F(x, y) dV + \dots \quad (3.11)$$

Each curved polygon  $\mathcal{P}$  is defined by its boundary, a piecewise smooth parametric curve:

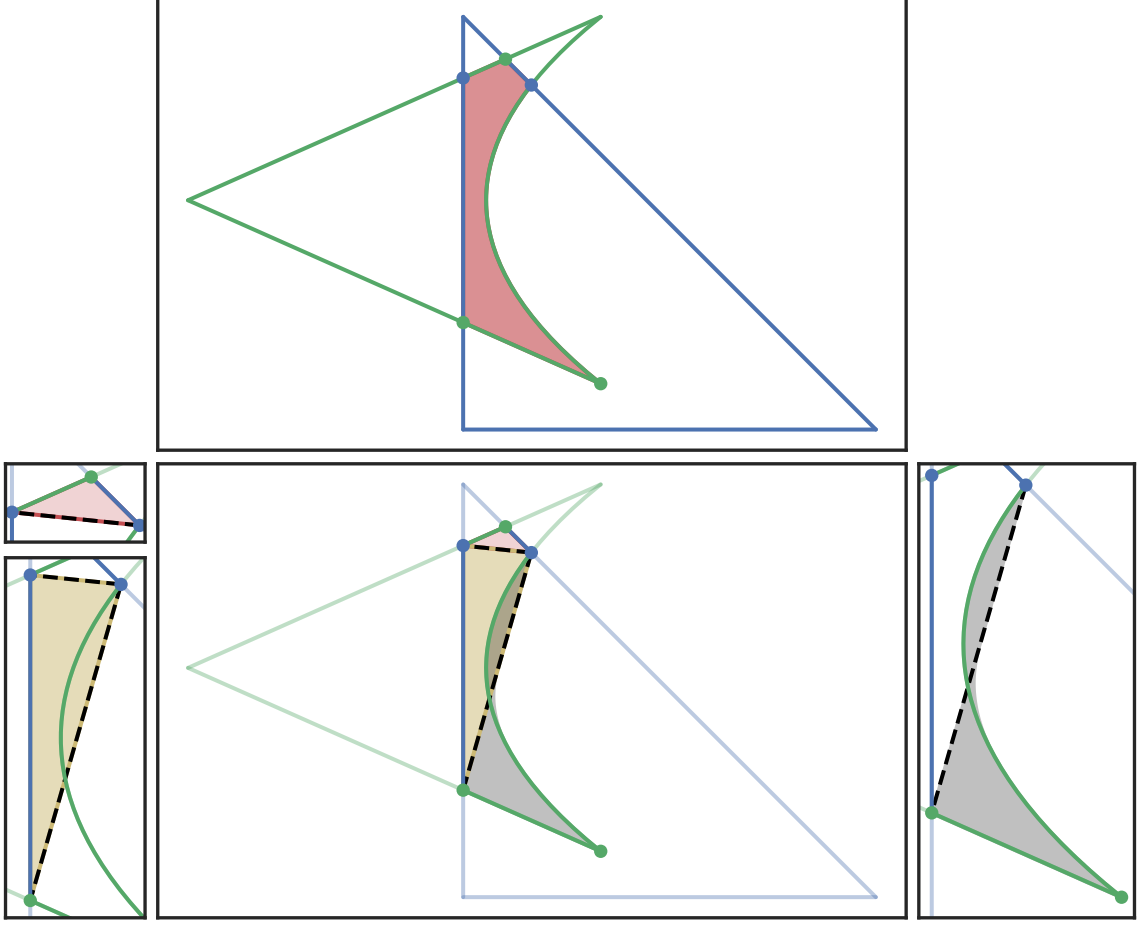
$$\partial\mathcal{P} = C_1 \cup \dots \cup C_n. \quad (3.12)$$

This can be thought of as a polygon with  $n$  sides that happens to have curved edges.

Quadrature rules for straight sided polygons have been studied [MXS09] though they are not in wide use. Even if a polygonal quadrature rule was to be employed, a map would need to be established from a reference polygon onto the curved edges. This map could then be used with a change of coordinates to move the integral from the curved polygon to the reference polygon. The problem of extending a mapping from a boundary to an entire domain has been studied as transfinite interpolation ([Che74, GT82, Per98]), barycentric coordinates ([Wac75]) or mean value coordinates ([Flo03]). However, these maps aren't typically suitable for numerical integration because they are either not bijective or increase the degree of the edges.

Since simple and well established quadrature rules do exist for triangles, a valid approach would be to tessellate a curved polygon into valid Bézier triangles (Figure 3.6) and then use substitution as in (3.7). However, tessellating is challenging for both theoretical and computational reasons. Theoretically, it's not even clear if an arbitrary curved polygon *can* be tessellated into Bézier triangles. Computationally, the placement of diagonals and potential introduction of interior nodes is very complex in cases where the curved polygon is nonconvex. What's more, the curved polygon is given only by the boundary, so higher-degree triangles (i.e., cubic and above) introduced during tessellation would need to place interior control points without causing the triangle to invert. To get a sense for these challenges, note how the “simple” introduction of diagonals in Figure 3.7 leads to one inverted element (the gray element) and another element with area outside of the curved polygon (the yellow element). Inverted Bézier triangles are problematic because the accompanying mapping leaves the boundary established by the edge curves. For example, if the tessellation of  $\mathcal{P}$  contains an inverted Bézier triangle  $\mathcal{T}_2 = b(\mathcal{U})$  then we'll need to numerically integrate

$$\int_{\mathcal{T}_2} \phi_0 \phi_1 dx dy = \int_{\mathcal{U}} |\det(Db)| (\phi_0 \circ b) (\phi_1 \circ b) ds dt. \quad (3.13)$$



**Figure 3.7.** Curved polygon intersection that can't be tessellated (with valid Bézier triangles) by introducing diagonals.

If the shape functions are from the preimage basis (see Section 2.5), then  $\phi_0$  will not be defined at  $b(s, t) \notin \mathcal{T}_0$  (similarly for  $\phi_1$ ). Additionally, the absolute value in  $|\det(Db)|$  makes the integrand non-smooth since for inverted elements  $\det(Db)$  takes both signs. If the shape functions are from the global coordinates basis, then tessellation can be used via an application of Theorem B.1, however this involves more computation than just applying Green's theorem along  $\partial\mathcal{P}$ . By applying the theorem, inverted elements may be used in a tessellation, for example, by introducing artificial diagonals from a vertex as in Figure 3.6. In the global coordinates basis,  $F = \phi_0\phi_1$  can be evaluated for points in an inverted element that leave  $\mathcal{T}_0$  or  $\mathcal{T}_1$ .

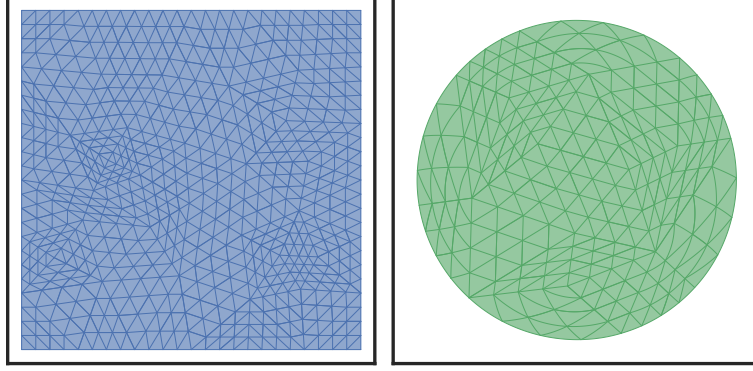
Instead, we focus on a Green's theorem based approach. Define horizontal and vertical antiderivatives  $H, V$  of the integrand  $F$  such that  $H_x = V_y = F$ . We make these *unique* by imposing the extra condition that  $H(0, y) \equiv V(x, 0) \equiv 0$ . This distinction is arbitrary, but in order to *evaluate*  $H$  and  $V$ , the univariate functions  $H(0, y)$  and  $V(x, 0)$  must be specified. Green's theorem tells us that

$$\int_{\mathcal{P}} 2F dV = \int_{\mathcal{P}} H_x + V_y dV = \oint_{\partial\mathcal{P}} H dy - V dx = \sum_j \int_{C_j} H dy - V dx. \quad (3.14)$$

For a given curve  $C$  with components  $x(r), y(r)$  defined on the unit interval, this amounts to having to integrate

$$G(r) = H(x(r), y(r))y'(r) - V(x(r), y(r))x'(r). \quad (3.15)$$

To do this, we'll use Gaussian quadrature with degree of accuracy sufficient to cover the degree of  $G(r)$ .



**Figure 4.1.** Some example meshes used during the numerical experiments; the donor mesh is on the left (blue) and the target mesh is on the right (green). These meshes are the cubic approximations of each domain and have been refined twice.

If the shape functions are in the global coordinates basis (Section 2.5), then  $G$  will also be a polynomial. This is because for these shape functions  $F = \phi_0\phi_1$  will be polynomial on  $\mathbf{R}^2$  and so will  $H$  and  $V$  and since each curve  $C$  is a Bézier curve segment, the components are also polynomials. If the shape functions are in the preimage basis, then  $F$  won't in general be polynomial; hence standard quadrature rules can't be exact.

To evaluate  $G$ , we must also evaluate  $H$  and  $V$  numerically. For example, since  $H(0, y) \equiv 0$ , the fundamental theorem of calculus tells us that  $H(\alpha, \beta) = \int_0^\alpha F(x, \beta) dx$ . To compute this integral via Gaussian quadrature

$$H(\alpha, \beta) \approx \frac{\alpha}{2} \sum_j w_j F\left(\frac{\alpha}{2}(x_j + 1), \beta\right) \quad (3.16)$$

we must be able to evaluate  $F$  for points on the line  $y = \beta$  for  $x$  between 0 and  $\alpha$ . If the shape functions are from the preimage basis, it may not even be possible to evaluate  $F$  for such points. Since  $\begin{bmatrix} \alpha & \beta \end{bmatrix}^T$  is on the boundary of  $\mathcal{P}$ , without loss of generality assume it is on the boundary of  $\mathcal{T}_0$ . Thus, for some elements (e.g., if the point is on the bottom of the element), points  $\begin{bmatrix} \nu & \beta \end{bmatrix}^T$  may not be in  $\mathcal{T}_0$ . Since  $\text{supp}(\phi_0) = \mathcal{T}_0$ , we could take  $\phi_0(\nu, \beta) = 0$ , but this would make the integrand non-smooth and so the accuracy of the *exact* quadrature would be lost. But extending  $\phi_0 = \hat{\phi}_0 \circ b_0^{-1}$  outside of  $\mathcal{T}_0$  may be impossible: even though  $b_0$  is bijective on  $\mathcal{U}$  it may be many-to-one elsewhere; hence  $b_0^{-1}$  can't be reliably extended outside of  $\mathcal{T}_0$ .

Even when the shape functions are from the global coordinates basis, setting  $H(0, y) \equiv 0$  may introduce quadrature points that are very far from  $\mathcal{P}$ . This can be somewhat addressed by using  $H(m, y) \equiv 0$  for a suitably chosen  $m$  (for instance, the minimum  $x$ -value in  $\mathcal{P}$ ). Then we have  $H(\alpha, \beta) = \int_m^\alpha F(x, \beta) dx$ .

#### 4. NUMERICAL EXPERIMENTS

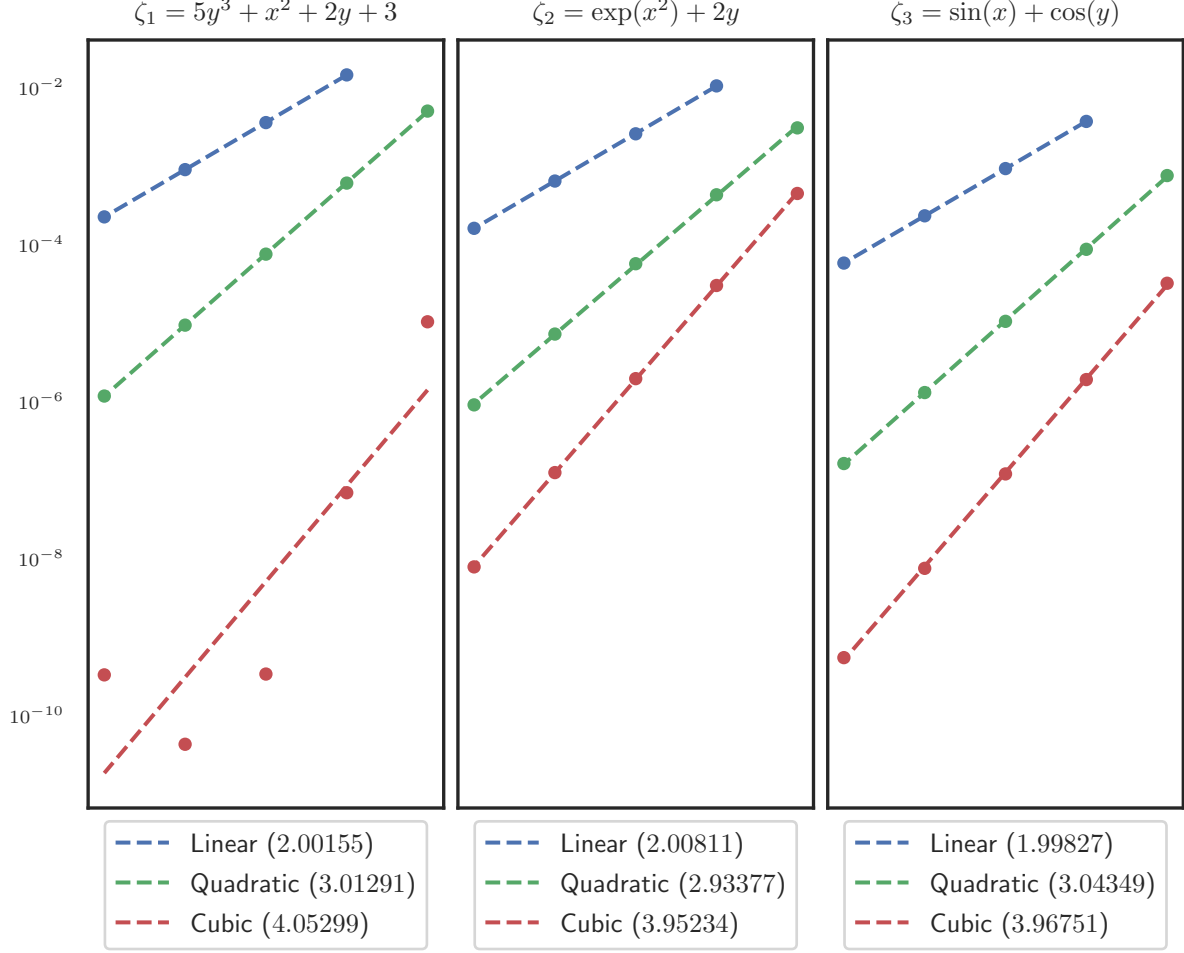
A numerical experiment was performed to investigate the observed order of convergence. Three pairs of random meshes were generated, the donor on a square of width  $17/8$  centered at the origin and the target on the unit disc. These domains were chosen intentionally so that the target mesh was completely covered by the donor mesh and the boundaries did not accidentally introduce ill-conditioned intersection between elements. The pairs were linear, quadratic and cubic approximations of the domains. The convergence test was done by refining each pair of meshes four times and performing solution transfer at each level. Figure 4.1 shows the cubic pair of meshes after two refinements.

Taking after [FM11], we transfer three discrete fields in the discontinuous Galerkin (DG) basis. Each field is derived from one of the smooth scalar functions

$$\zeta_1(x, y) = 5y^3 + x^2 + 2y + 3, \quad (4.1)$$

$$\zeta_2(x, y) = \exp(x^2) + 2y, \quad (4.2)$$

$$\zeta_3(x, y) = \sin(x) + \cos(y). \quad (4.3)$$



**Figure 4.2.** Convergence results for scalar fields on three pairs of related meshes: a linear, quadratic and cubic mesh of the same domain.

For a given mesh size  $h$ , we expect that on a degree- $p$  isoparametric mesh our solution transfer will have  $\mathcal{O}(h^{p+1})$  errors. To measure the rate of convergence:

- Choose a mesh pair  $\mathcal{M}_D, \mathcal{M}_T$  and a known function  $\zeta(x, y)$ .
- Refine the meshes recursively, starting with  $\mathcal{M}_D^{(0)} = \mathcal{M}_D, \mathcal{M}_T^{(0)} = \mathcal{M}_T$ .
- Create meshes  $\mathcal{M}_D^{(j)}$  and  $\mathcal{M}_T^{(j)}$  from  $\mathcal{M}_D^{(j-1)}$  and  $\mathcal{M}_T^{(j-1)}$  by subdividing each curved element into four elements.
- Approximate  $\zeta$  by a discrete field: the nodal interpolant on the donor mesh  $\mathcal{M}_D^{(j)}$ . The nodal interpolant is constructed by evaluating  $\zeta$  at the nodes  $\mathbf{n}_j$  corresponding to each shape function:

$$\mathbf{f}_j = \sum_i \zeta(\mathbf{n}_i) \phi_D^{(i)}. \quad (4.4)$$

- Transfer  $\mathbf{f}_j$  to the discrete field  $\mathbf{g}_j$  on the target mesh  $\mathcal{M}_T^{(j)}$ .
- Compute the relative error on  $\mathcal{M}_T^{(j)}$ :  $E_j = \|\mathbf{g}_j - \zeta\|_2 / \|\zeta\|_2$  (here  $\|\cdot\|_2$  is the  $L_2$  norm on the target mesh).

We should instead be measuring  $\|\mathbf{g}_j - \mathbf{f}_j\|_2 / \|\mathbf{f}_j\|_2$ , but  $E_j$  is much easier to compute for  $\zeta(x, y)$  that are straightforward to evaluate. Due to the triangle inequality  $E_j$  can be a reliable proxy for the actual projection error, though when  $\|\mathbf{f}_j - \zeta\|_2$  becomes too large it will dominate the error and no convergence will be observed.

Convergence results are shown in Figure 4.2 and confirm the expected orders. Since  $\zeta_1$  is a cubic polynomial, one might expect the solution transfer on the cubic mesh to be *exact*. This expectation would hold in the superparametric case (i.e., where the elements are straight sided but the shape functions are higher degree). However, when the elements aren't straight sided the isoparametric mapping changes the underlying function space.

## 5. CONCLUSION

We have described a method for conservative interpolation between curved meshes. The transfer process conserves globally to machine precision since we can use exact quadratures for all integrals. The primary source of error comes from solving the linear system with the mass matrix for the target mesh. This allows less restrictive usage of mesh adaptivity, which can make computations more efficient. Additionally, having a global transfer algorithm allows for remeshing to be done less frequently.

The algorithm breaks down into three core subproblems: Bézier triangle intersection, an expanding front for intersecting elements and integration on curved polygons. The inherently local nature of the expanding front allows the algorithm to be parallelized via domain decomposition with little data shared between processes. By restricting integration to the intersection of elements from the target and donor meshes, the algorithm can accurately transfer both continuous and discontinuous fields.

As mentioned in the preceding chapters, there are several research directions possible to build upon the solution transfer algorithm. The usage of Green's theorem nicely extends to  $\mathbf{R}^3$  via Stokes' theorem, but the Bézier triangle intersection algorithm is specific to  $\mathbf{R}^2$ . The equivalent Bézier tetrahedron intersection algorithm is significantly more challenging.

The restriction to shape functions from the global coordinates basis is a symptom of the method and not of the inherent problem. The preimage basis has several appealing properties, for example, this basis can be precomputed on  $\mathcal{U}$ . The problem of a valid tessellation of a curved polygon warrants more exploration. Such a tessellation algorithm would enable usage of the preimage basis.

The usage of the global coordinates basis does have some benefits. In particular, the product of shape functions from different meshes is still a polynomial in  $\mathbf{R}^2$ . This means that we could compute the coefficients of  $F = \phi_0\phi_1$  directly and use them to evaluate the antiderivatives  $H$  and  $V$  rather than using the fundamental theorem of calculus. Even if this did not save any computation, it may still be preferred over the FTC approach because it would remove the usage of quadrature points outside of the domain  $\mathcal{P}$ .

## APPENDIX A. BÉZIER INTERSECTION PROBLEMS

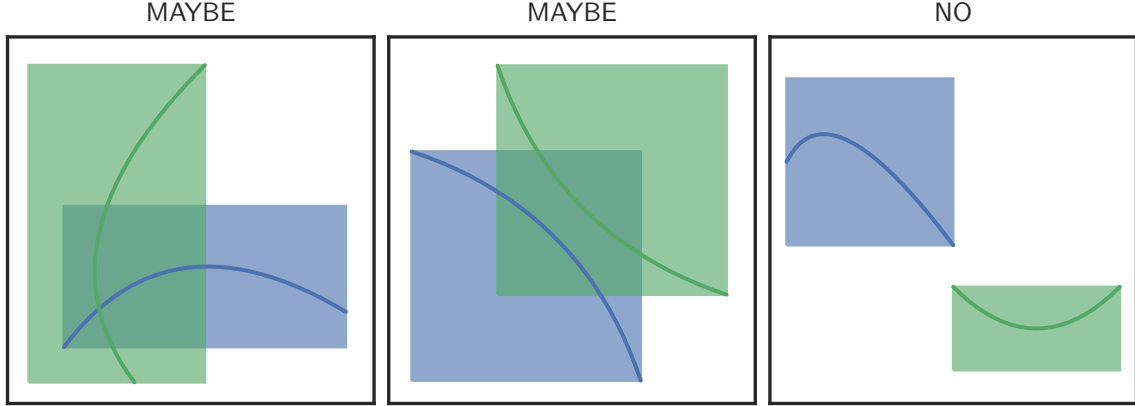
**A.1. Intersecting Bézier Curves.** The problem of intersecting two Bézier curves is a core building block for intersecting two Bézier triangles in  $\mathbf{R}^2$ . Since a curve is an algebraic variety of dimension one, the intersections will either be a curve segment common to both curves (if they coincide) or a finite set of points (i.e., dimension zero). Many algorithms, both geometric [SP86, SN90, KLS98] and algebraic [MD92], have been described in the literature.

In our implementation, the Bézier subdivision algorithm is used. In the case of a transversal intersection (that is, one where the tangents to each curve are not parallel and both are non-zero), this algorithm performs very well. However, when curves are tangent, a large number of (false) candidate intersections are detected and convergence of Newton's method slows once in a neighborhood of an actual intersection. Nontransversal intersections have infinite condition number, but transversal intersections with very high condition number can also cause convergence problems.

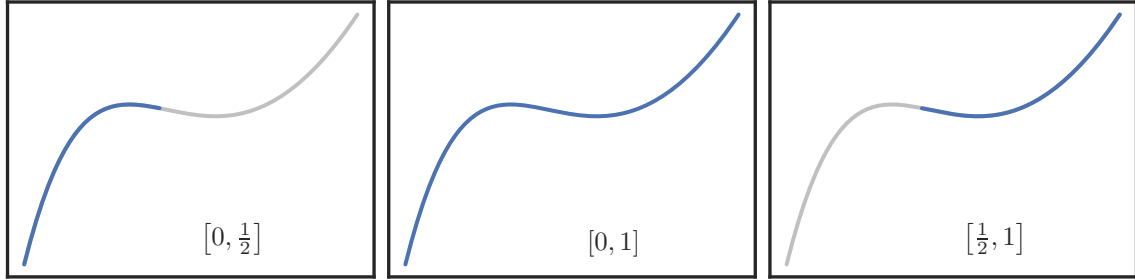
In the Bézier subdivision algorithm, we first check if the bounding boxes for the curves are disjoint (Figure A.1). We use the bounding boxes rather than the convex hulls since they are easier to compute and the intersections of boxes are easier to check. If they are disjoint, the pair can be rejected. If not, each curve  $\mathcal{C} = b([0, 1])$  is split into two halves by splitting the unit interval:  $b([0, \frac{1}{2}])$  and  $b([\frac{1}{2}, 1])$  (Figure A.2).

As the subdivision continues, some pairs of curve segments may be kept around that won't lead to an intersection (Figure A.3). Once the curve segments are close to linear within a given tolerance (Figure A.4), the process terminates.

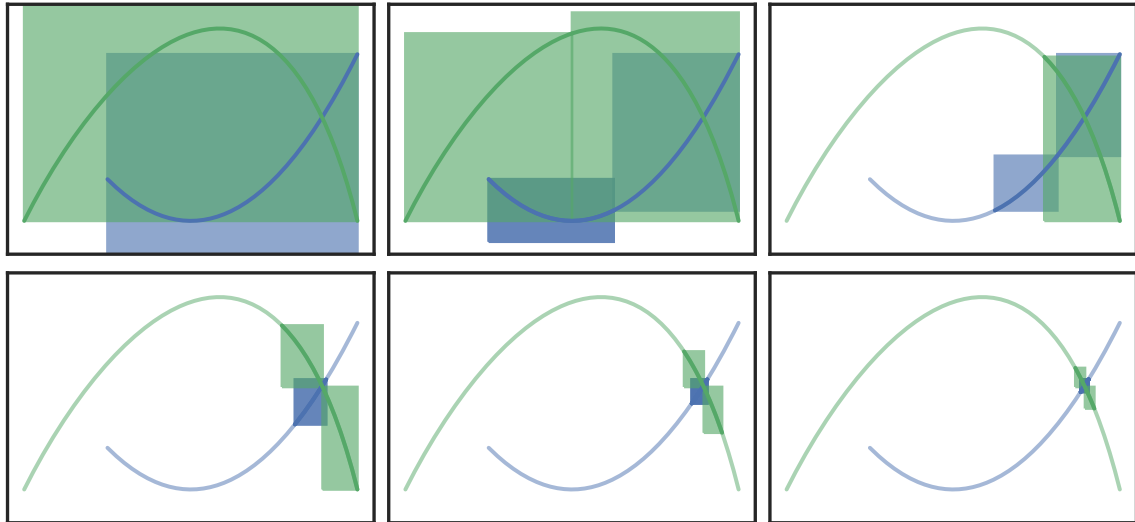
Once both curve segments are linear (to tolerance), the intersection is approximated by intersecting the lines connecting the endpoints of each curve segment. This approximation is used as a starting point for Newton's method, to find a root of  $F(s, t) = b_0(s) - b_1(t)$ . Since  $b_0(s), b_1(t) \in \mathbf{R}^2$  we have Jacobian



**Figure A.1.** Bounding box intersection predicate. This is a cheap way to conclude that two curves don't intersect, though it inherently is susceptible to false positives.



**Figure A.2.** Bézier curve subdivision.

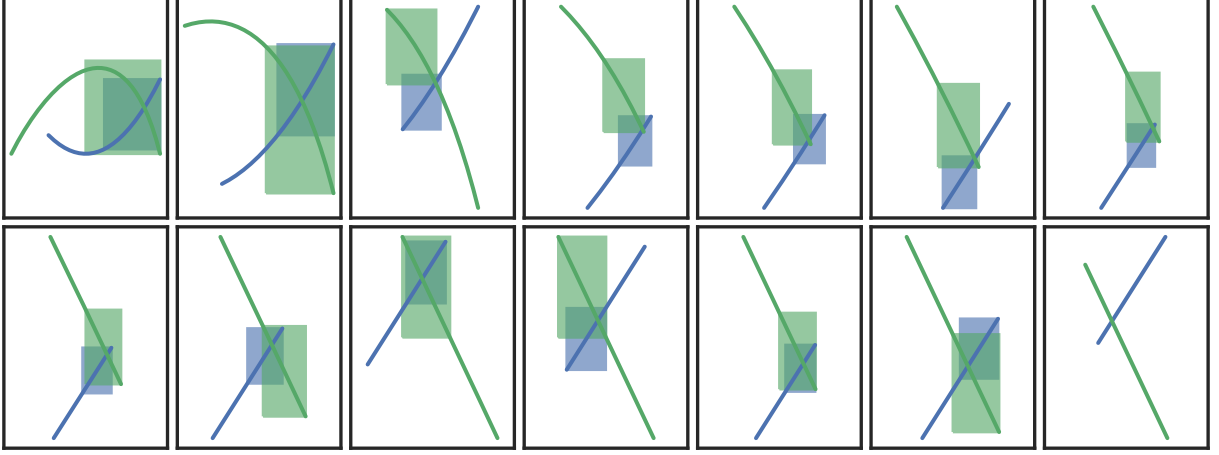


**Figure A.3.** Bézier subdivision algorithm.

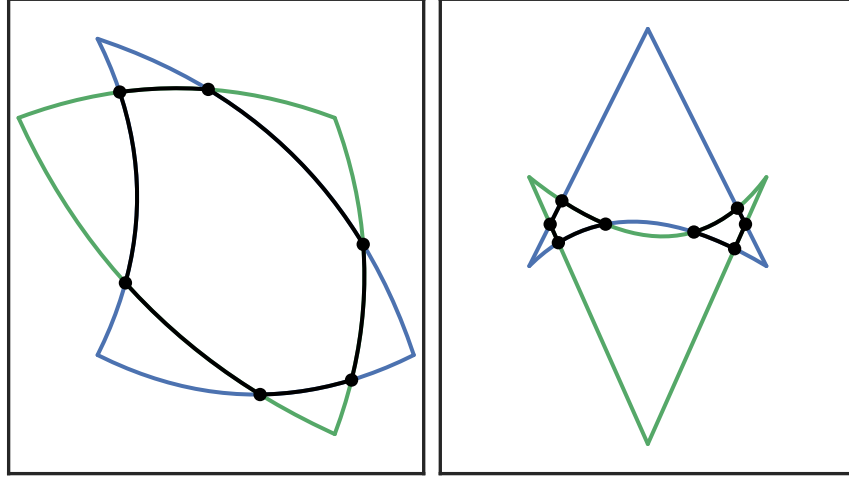
$J = \begin{bmatrix} b'_0(s) & -b'_1(t) \end{bmatrix}$ . With these, Newton's method is

$$\begin{bmatrix} s_{n+1} & t_{n+1} \end{bmatrix}^T = \begin{bmatrix} s_n & t_n \end{bmatrix}^T - J_n^{-1} F_n. \quad (\text{A.1})$$





**Figure A.4.** Subdividing until linear within tolerance.



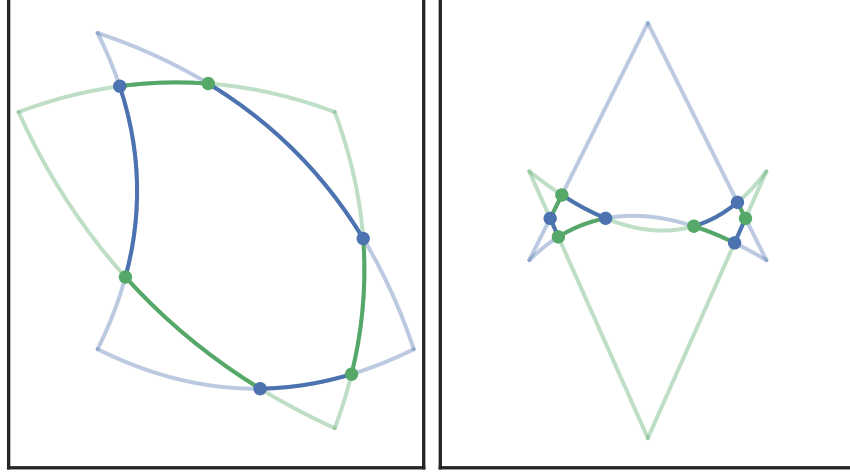
**Figure A.5.** Edge intersections during Bézier triangle intersection.

This also gives an indication why convergence issues occur at nontransversal intersections: they are exactly the intersections where the Jacobian is singular.

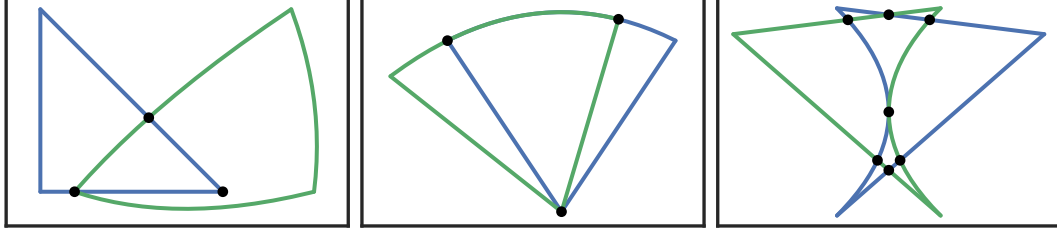
**A.2. Intersecting Bézier Triangles.** The chief difficulty in intersecting two surfaces is intersecting their edges, which are Bézier curves. Though this is just a part of the overall algorithm, it proved to be the *most difficult* to implement ([Her17]). So the first part of the algorithm is to find all points where the edges intersect (Figure A.5).

To determine the curve segments that bound the curved polygon region(s) (see Section 2.6 for more about curved polygons) of intersection, we not only need to keep track of the coordinates of intersection, we also need to keep note of *which* edges the intersection occurred on and the parameters along each curve. With this information, we can classify each point of intersection according to which of the two curves forms the boundary of the curved polygon (Figure A.6). Using the right-hand rule we can compare the tangent vectors on each curve to determine which one is on the interior.

This classification becomes more difficult when the curves are tangent at an intersection, when the intersection occurs at a corner of one of the surfaces or when two intersecting edges are coincident on the same algebraic curve (Figure A.7).



**Figure A.6.** Classified intersections during Bézier triangle intersection.



**Figure A.7.** Bézier triangle intersection difficulties.

In the case of tangency, the intersection is nontransversal, and hence has infinite condition number. In the case of coincident curves, there are infinitely many intersections (along the segment when the curves coincide) so the subdivision process breaks down.

In all of our experiments, we did not encounter any of these difficulties with our implementation. However, for fail-safe code, additional fallback strategies would be required, and we refer the reader to the extensive literature for further details on covering all possible special cases that might occur [SP86, SN90, KLS98].

A.2.1. *Example.* Consider two Bézier surfaces (Figure A.8)

$$b_0(s, t) = \begin{bmatrix} 8s \\ 8t \end{bmatrix} \quad \text{and} \quad b_1(s, t) = \begin{bmatrix} 2(6s + t - 1) \\ 2(8s^2 + 8st - 8s + 3t + 2) \end{bmatrix}. \quad (\text{A.2})$$

In the *first step* we find all intersections of the edge curves

$$\begin{aligned} E_0(r) &= \begin{bmatrix} 8r \\ 0 \end{bmatrix}, E_1(r) = \begin{bmatrix} 8(1-r) \\ 8r \end{bmatrix}, E_2(r) = \begin{bmatrix} 0 \\ 8(1-r) \end{bmatrix}, \\ E_3(r) &= \begin{bmatrix} 2(6r-1) \\ 4(2r-1)^2 \end{bmatrix}, E_4(r) = \begin{bmatrix} 10(1-r) \\ 2(3r+2) \end{bmatrix}, E_5(r) = \begin{bmatrix} -2r \\ 2(5-3r) \end{bmatrix}. \end{aligned} \quad (\text{A.3})$$

We find three intersections and we classify each of them by comparing the tangent vectors

$$I_1 : E_2 \left( \frac{7}{9} \right) = E_3 \left( \frac{1}{6} \right) = \frac{16}{9} \begin{bmatrix} 0 \\ 1 \end{bmatrix} \implies E_2' \left( \frac{7}{9} \right) \times E_3' \left( \frac{1}{6} \right) = 96, \quad (\text{A.4})$$

$$I_2 : E_0 \left( \frac{1}{2} \right) = E_3 \left( \frac{1}{2} \right) = \begin{bmatrix} 4 \\ 0 \end{bmatrix} \implies E_0' \left( \frac{1}{2} \right) \times E_3' \left( \frac{1}{2} \right) = 0, \quad (\text{A.5})$$

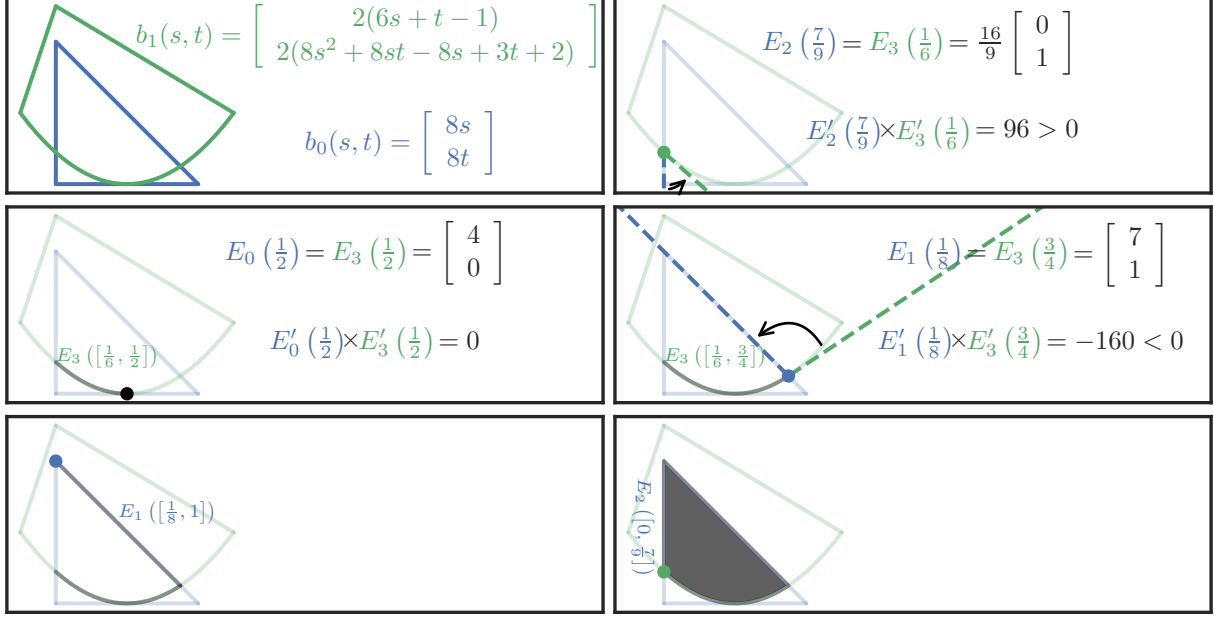


Figure A.8. Surface Intersection Example

$$I_3 : E_1\left(\frac{1}{8}\right) = E_3\left(\frac{3}{4}\right) = \begin{bmatrix} 7 \\ 1 \end{bmatrix} \Rightarrow E'_1\left(\frac{1}{8}\right) \times E'_3\left(\frac{3}{4}\right) = -160. \quad (\text{A.6})$$

From here, we construct our curved polygon intersection by drawing from our list of intersections until none remain.

- First consider  $I_1$ . Since  $E'_2 \times E'_3 > 0$  at this point, then we consider the curve  $E_3$  to be *interior*.
- After classification, we move along  $E_3$  until we encounter another intersection:  $I_2$ .
- $I_2$  is a point of tangency since  $E'_0(\frac{1}{2}) \times E'_3(\frac{1}{2}) = 0$ . Since a tangency has no impact on the underlying intersection geometry, we ignore it and keep moving.
- Continuing to move along  $E_3$ , we encounter another intersection:  $I_3$ . Since  $E'_1 \times E'_3 < 0$  at this point, we consider the curve  $E_1$  to be *interior* at the intersection. Thus we stop moving along  $E_3$  and we have our first curved segment:  $E_3\left(\left[\frac{1}{6}, \frac{3}{4}\right]\right)$ .
- Finding no other intersections on  $E_1$  we continue until the end of the edge. Now our (ordered) curved segments are

$$E_3\left(\left[\frac{1}{6}, \frac{3}{4}\right]\right) \rightarrow E_1\left(\left[\frac{1}{8}, 1\right]\right). \quad (\text{A.7})$$

- Next we stay at the corner and switch to the next curve  $E_2$ , moving along that curve until we hit the next intersection  $I_1$ . Now our (ordered) curved segments are

$$E_3\left(\left[\frac{1}{6}, \frac{3}{4}\right]\right) \rightarrow E_1\left(\left[\frac{1}{8}, 1\right]\right) \rightarrow E_2\left(\left[0, \frac{7}{9}\right]\right). \quad (\text{A.8})$$

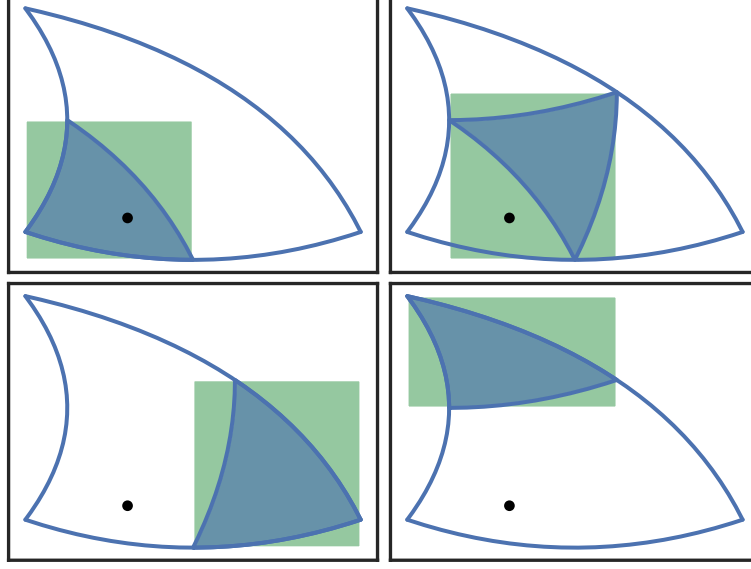
Since we are now back where we started (at  $I_1$ ) the process stops.

We represent the boundary of the curved polygon as Bézier curves, so to complete the process we reparameterize [Far01, Ch. 5.4] each curve onto the relevant interval. For example,  $E_3$  has control points  $p_0 = \begin{bmatrix} -2 \\ 4 \end{bmatrix}$ ,

$p_1 = \begin{bmatrix} 4 \\ -4 \end{bmatrix}$ ,  $p_2 = \begin{bmatrix} 10 \\ 4 \end{bmatrix}$  and we reparameterize on  $\alpha = \frac{1}{6}, \beta = \frac{3}{4}$  to control points

$$q_0 = E_3\left(\frac{1}{6}\right) = \frac{16}{9} \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad (\text{A.9})$$

$$q_1 = (1 - \alpha)((1 - \beta)p_0 + \beta p_1) + \alpha((1 - \beta)p_1 + \beta p_2) = \frac{1}{6} \begin{bmatrix} 21 \\ -8 \end{bmatrix}, \quad (\text{A.10})$$



**Figure A.9.** Checking for a point  $\mathbf{p}$  in each of four subregions when subdividing a Bézier triangle.

$$q_2 = E_3 \left( \frac{3}{4} \right) = \begin{bmatrix} 7 \\ 1 \end{bmatrix}. \quad (\text{A.11})$$

**A.3. Bézier Triangle Inverse.** The problem of determining the parameters  $(s, t)$  given a point  $\mathbf{p} = \begin{bmatrix} x & y \end{bmatrix}^T$  in a Bézier triangle can also be solved by using subdivision with a bounding box predicate and then Newton's method at the end.

For example, Figure A.9 shows how regions of  $\mathcal{U}$  can be discarded recursively until the suitable region for  $(s, t)$  has a sufficiently small area. At this point, we can apply Newton's method to the map  $F(s, t) = b(s, t) - \mathbf{p}$ . It's very helpful (for Newton's method) that  $F : \mathbf{R}^2 \rightarrow \mathbf{R}^2$  since the Jacobian will always be invertible when the Bézier triangle is valid. If  $\mathbf{p} \in \mathbf{R}^3$  then the system would be underdetermined. Similarly, if  $\mathbf{p} \in \mathbf{R}^2$  but  $b(s)$  is a Bézier curve then the system would be overdetermined.

## APPENDIX B. ALLOWING TESSELLATION WITH INVERTED TRIANGLES

**Theorem B.1.** Consider three smooth curves  $b_0, b_1, b_2$  that form a closed loop:  $b_0(1) = b_1(0)$ ,  $b_1(1) = b_2(0)$  and  $b_2(1) = b_0(0)$ . Take *any* smooth map  $\varphi(s, t)$  on  $\mathcal{U}$  that sends the edges to the three curves:

$$\varphi(r, 0) = b_0(r), \quad \varphi(1-r, r) = b_1(r), \quad \varphi(0, 1-r) = b_2(r) \quad \text{for } r \in [0, 1]. \quad (\text{B.1})$$

Then we must have

$$2 \int_{\mathcal{U}} \det(D\varphi) [F \circ \varphi] dt ds = \oint_{b_0 \cup b_1 \cup b_2} H dy - V dx \quad (\text{B.2})$$

for antiderivatives that satisfy  $H_x = V_y = F$ .

When  $\det(D\varphi) > 0$ , this is just the change of variables formula combined with Green's theorem.

*Proof.* Let  $x(s, t)$  and  $y(s, t)$  be the components of  $\varphi$ . Define

$$\Delta S = H(x, y)y_s - V(x, y)x_s \quad \text{and} \quad \Delta T = H(x, y)y_t - V(x, y)x_t. \quad (\text{B.3})$$

On the unit triangle  $\mathcal{U}$ , Green's theorem gives

$$\int_{\mathcal{U}} [\partial_s \Delta T - \partial_t \Delta S] dV = \oint_{\partial \mathcal{U}} \Delta S ds + \Delta T dt. \quad (\text{B.4})$$

The boundary  $\partial \mathcal{U}$  splits into the bottom edge  $E_0$ , hypotenuse  $E_1$  and left edge  $E_2$ .

Since

$$E_0 = \left\{ \begin{bmatrix} r \\ 0 \end{bmatrix} \mid r \in [0, 1] \right\} \quad (\text{B.5})$$

we take  $\varphi(r, 0) = b_0(r)$ ; hence

$$dx = x_s dr, dy = y_s dr \implies H dx - V dy = \Delta S dr. \quad (\text{B.6})$$

We also have  $ds = dr$  and  $dt = 0$  due to the parameterization; thus

$$\int_{E_0} \Delta S ds + \Delta T dt = \int_{r=0}^{r=1} \Delta S dr = \int_{b_0} H dx - V dy. \quad (\text{B.7})$$

We can similarly verify that  $\int_{E_j} \Delta S ds + \Delta T dt = \int_{b_j} H dx - V dy$  for the other two edges. Combining this with (B.4) we have

$$\int_{\mathcal{U}} [\partial_s \Delta T - \partial_t \Delta S] dV = \oint_{b_0 \cup b_1 \cup b_2} H dx - V dy. \quad (\text{B.8})$$

To complete the proof, we need

$$\int_{\mathcal{U}} [\partial_s \Delta T - \partial_t \Delta S] dV = 2 \int_{\mathcal{U}} \det(D\varphi) [F \circ \varphi] dV \quad (\text{B.9})$$

but one can show directly that

$$\partial_s \Delta T - \partial_t \Delta S = 2(x_s y_t - x_t y_s) F(x, y) = 2 \det(D\varphi) [F \circ \varphi]. \quad \blacksquare$$

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